

## CLAIMS

1. Method of keying, in a space presenting two spatial dimensions and one temporal dimension, a signal S measured in positions U subject to an uncertainty, from a set of N signals measured in determined positions, the N + 1 signals having their temporal origin in a same plane, the said method involving: - re-sampling the N + 1 signals in order to place them all in an identical sampling range,

- filtering the signal S in order to place it in a range of frequencies that is identical to that of the N signals,

and wherein the method also involves:

- defining for each position U associated with the measurements of the signal S a same neighbourhood of places V in the spatio-temporal space centred on the position U,

- producing a layered neural network  $RN_v$  for each location V in the neighbourhood of U, each network having an entry vector of dimension N associated with the measurements of the N signals and a scalar exit associated with a measurement of the signal S,

- for each neural network  $RN_v$ , defining a learning set such that the entries are the collection of all the vectors of measurements of the N signals situated at the locations V and the exits are the collection of the values of the signal S at the positions U for all the positions U,

- fixing a predetermined number of iterations  $N_{it}$  for all the neural networks and launching the learning phases of all the networks,

- for each neural network  $RN_v$ , calculating the value of the integral  $I_v$  of the function giving the error committed by the network at each iteration, from iteration 1 to iteration  $N_{it}$ .

- for each surface spatial position  $V_k$  of the neighbourhood with coordinates  $(X_k, Y_k, t_0)$ , selecting in the time dimension the pair of locations  $V_{1k}(X_k, Y_k, t_1)$ ,  $V_{2k}(X_k, Y_k, h)$ , of the neighbourhood which correspond to the two smallest local minima of the two integrals  $(LV, LV_{2k})$ ,

- for each surface spatial position  $V_k$  of the neighbourhood, retaining from among the two positions  $V_h(X_k, Y_k, t_1)$ ,  $V_{2k}(X_k, Y_k, h)$  the position  $V_m$ , for which the signal estimated by the respective neural networks  $RNV$  and  $RNV_{2k}$  presents a maximum variance,

- choosing from among the positions  $V_m$  the position  $V_{eal}$  for which the integral  $tV_m$  is minimum.

2. Method according to claim 1, wherein the use of the neural networks comprises:

- defining for each position  $U$  associated with the measurements of the signal  $S$  a same neighbourhood of places  $V$  in the spatio-temporal space centred on the position  $U$ ,

- producing a layered neural network  $RNV$  for each location  $V$  in the neighbourhood of  $U$ . each network having an entry vector of dimension  $N \times M$  associated with the measurements on a time window of size  $M$  centred on  $V$  of the  $N$  signals and a scalar exit associated with a value of the signal  $S$ ,

- for each neural network, defining a learning set such that the entries are the collection of all the vectors of measurements taken in a time window of size  $M$  centred on  $V$  for the  $N$  signals and the exits are the collection of the values of the signal  $S$  at the positions  $U$  for all the positions  $U$ ,

- fixing a predetermined number of iterations  $N_{it}$  for all the neural networks and launching the learning phases of all the networks,

- for each neural network  $RN_v$ , calculating the value of the integral  $I_v$  of the function giving the error committed by the network at each iteration, from iteration 1 to iteration  $N_{it}$ ,

- for each surface spatial position  $V_k$  of the neighbourhood with coordinates  $(X_k, Y_k, t_0)$ , selecting in the time dimension the pair of locations  $V_{1k}(X_k, Y_k, t_1)$ ,  $V_{2k}(X_k, Y_k, t_2)$ , of the neighbourhood which correspond to the two smallest local minima of the two integrals  $(I_{v1k}, I_{v2k})$ .

- for each surface spatial position  $V_k$  of the neighbourhood, retaining from among the two positions  $V_{1k}(X_k, Y_k, t_1)$ ,  $V_{2k}(X_k, Y_k, t_2)$  the position  $V_m$ , for which the signal estimated by the respective neural networks  $RN_{v1}$  and  $RN_{v2k}$  presents a maximum variance,

- choosing from among the  $V_m$  positions the position  $V_{eal}$  for which the integral  $I_{v_{eal}}$  is minimum.